

PREDICTION OF GREY WATER FOOTPRINT
OF SUNGAI LEMBING, BUKIT SAGU AND
BUKIT UBI WATER TREATMENT PLANTS

SITI FAZLINA BINTI MOHD SUHAIMI

B. ENG (HONS.) CIVIL ENGINEERING

UNIVERSITI MALAYSIA PAHANG



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(Student's Signature)

Full Name : SITI FAZLINA BINTI MOHD SUHAIMI

ID Number : AA15168

Date : 31st MAY 2019

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SITI FAZLINA BINTI MOHD SUHAIMI

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ABSTRAK

Faktor terpenting yang mempengaruhi kekurangan air di dalam dan luar negara dan ketersediaan sumber air tawar bukan sahaja penduduk dunia yang semakin berkembang tetapi juga peningkatan permintaan air. Dari kajian ini, tahap pencemaran air di lembah sungai Kuantan direkodkan mengikut setiap loji rawatan air (WTP) dan penilaian jejak air kelabu digunakan sebagai pendekatan untuk mengira jumlah air tawar yang digunakan untuk mengasimilasi kepekatan pencemar. Oleh itu, kajian ini bertujuan untuk mengira jejak air kelabu keseluruhan, untuk meramalkan trend keseluruhan jejak air kelabu dan membandingkan algoritma terbaik antara Rangkaian Neural Buatan (ANN) dan Bayesian Networks (BN) dalam ramalan jejak air kelabu di Sungai Lembing WTP, Bukit Sagu WTP dan Bukit Ubi WTP pada tahun 2015 hingga 2017. Sebagai hasil akhir kajian ini, jumlah air kelabu di Sungai Lembing, Bukit Sagu dan loji rawatan air Bukit Ubi di lembah sungai Kuantan dikira. Trend ramalan keseluruhan jejak air kelabu dalam tiga loji rawatan air telah dapat dihasilkan sebagai hasil akhir kajian. Algoritma Rangkaian Neural Buatan (ANN) juga dipilih sebagai algoritma terbaik.

ABSTRACT

The most important factors affecting water scarcity in local and global and the availability of fresh water resources are not only a growing world population but also an increasing water demand. From this study, the level pollution of water in Kuantan river basin is recorded according to each water treatment plant (WTP) and grey water footprint assessment was used as an approach to account the total amount of freshwater used to assimilate the pollutant's concentration. Hence, this study is aimed to calculate the total grey water footprint, to predict the trend of total grey water footprint and to compare the best algorithm between Artificial Neural Network (ANN) and Bayesian Networks (BN) in grey water footprint prediction at Sungai Lembing WTP, Bukit Sagu WTP and Bukit Ubi WTP in 2015 until 2017. As the end result of this study, the total grey water footprint in Sungai Lembing, Bukit Sagu and Bukit Ubi water treatment plant in Kuantan river basin is calculated. Prediction trend of total grey water footprint in three water treatment plants has able to be produced. Artificial Neural Network (ANN) algorithm is also be chosen as the best algorithm.

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LIST OF SYMBOLS

| | |
|------------------------|------------------------|
| Fe^2 | Iron |
| m^3/s | Meter cubic per second |
| L | Load of pollutant |
| $\text{NO}_3\text{-N}$ | Nitrate-nitrogen |

LIST OF ABBREVIATIONS

| | |
|--------|--|
| ANN | Artificial Neural Network |
| BN | Bayesian Networks |
| BOD | Biochemical Oxygen Demand |
| COD | Chemical Oxygen Demand |
| LOCF | Last Observation Carried Forward |
| JPS | Jabatan Pengairan dan Saliran Pahang |
| MMD | Malaysian Meteorological Department |
| PAIP | Pengurusan Air Pahang Berhad |
| RMSE | Root Mean Square Error |
| WF | Water Footprint |
| WFgrey | Grey Water Footprint |
| WTP | Water Treatment Plants |
| WEKA | Wakaito Environment for Knowledge Analysis |

CHAPTER 1

INTRODUCTION

1.1 Background of Study

Water is most important substance for all living things include plants and animals to survive on the earth. People depend on water not only for drinking but also cooking, washing, carrying away wastes and other domestic needs. Water is an important factor of production contributing both directly and indirectly to economic activity across all sectors and regions of the global economy (Distefano & Kelly, 2017). Potable water or water that is safe for drinking must be free of germs and chemicals and be clear because water is a good carrier of disease germs. When the water becomes non-potable or contaminated, people can get serious illnesses if they keep use this polluted water. Diseases that produce bacteria, toxic substances and excessive amounts of minerals and organic matter should be avoided and overcome because the water used by the public must be clean and safe. Therefore, water purification works are very important to ensure that all impurities and bacteria from the air are removed and make it healthy. Water supply systems must also meet requirements for public, commercial, and industrial activities. In all cases, the water must fulfil both quality and quantity requirements.

Water supply system gets water from an assortment of areas after proper treatment; including groundwater (aquifers), surface water, for example, lakes and streams, and the ocean through desalination. Water treatment plant is the process of converting raw water which the water is taken from the river to clean water supply to the residence areas. Raw water is full of contaminants including bacteria, chemicals and other toxins. Its treatment aims to reduce the contaminants to acceptable levels to make the water safe for people to use. Therefore, water treatment plant is needed to treat raw

water. The overview of water treatment plant is first take the raw water from river and it is treated with alum, lime, fluoride, chlorine. Then it has to undergo the process of filtration, send to the master station and remove the sludge in the water before distributed to the consumers.

The most important factors affecting water scarcity in local and global availability of fresh water resources not only pollution and climate change but also a growing world population and an increasing water demand. The increasing demand for fresh water is the main challenge to sustain water utilization all over the world. Freshwater represents to 2.5% of Earth's water and is progressively threatened by human activity and climate change (Distefano & Kelly, 2017). Water has been largely studied by engineers in Jordan, while little research has adopted a discourse analysis procedure to water scarcity in the country (Hussein, 2018). While the concept of water scarcity is generally current topic, it is the difficulty of getting sources of clean water for utilize amid a period of time and may result in encourages reduction and disintegration of accessible water resources. The total energy and water use in China has been clearly increasing in the last decades as China has been experiencing a dramatic economic development (Xu, Li, & Lu, 2017).

An increasing number of studies have been carried out since in 1990s, to quantify and to investigate the existing differences between water demand, water supply and the geographical distance between them, the concepts of water footprint and virtual water trade (Arto, Andreoni, & Rueda-Cantuche, 2016). The water footprint (WF) of a product or process was introduced for the first time in 2003 and is defined as the volume of freshwater consumed and polluted to produce a product. The water footprint is further analysed in three parts: the blue, green and grey water footprints. The blue WF is an indicator of the surface water or groundwater consumption, which includes the evaporated water, incorporated into the product, and lost return flow. The green WF is defined as the consumption of water from precipitation that is stored in the soil and does not run off or recharge the ground-water and thus, is available for evapotranspiration of plants. Finally, the grey WF of a process step indicates the degree of freshwater pollution that can be associated with the process step. The grey WF is defined as the volume of freshwater that is required to assimilate the load of pollutants

based on natural background concentrations and existing ambient water quality standards (A Y Hoekstra & Mekonnen, 2011).

1.2 Problem Statement

It is necessary to have a knowledge of water resource consumption and pollution during the life cycle of energy production to reduce water consumption (Ding, Liu, Yang, & Lu, 2018). Water treatment plants at Sungai Lembing, Bukit Sagu and Bukit Ubi in Kuantan River Basin have enough water quantity but using ineffective long-term the management. Inadequate infrastructure and resources will bring problems in handling wastewater management efficiently and sustainably for the majority of cities in developing countries (Ding et al., 2018). According (Cha, Son, Hong, An, & Part, 2017), one of measures against water depletion and degradation by human activities by applied the efficient water management. In addition, the data for overall water which includes rain and evaporation also is not recorded. The data are important for any researches as this data can give a solution to any problems that comes from the water.

An increasing in demand for water and a decrease in availability and quality will affect the freshwater scarcity and pollution will be aggravated problems in the future (Ercin & Hoekstra, 2014). The Intergovernmental Panel on Climate Change (IPCC) reported that unabated climate change has the potential to strongly impact freshwater resources with wide ranging consequences for societies and ecosystems (Murray, Foster, & Prentice, 2012).

Water Footprints can be analysed as blue, green or grey water footprints. According to (Zhi, Yang, Yin, Hamilton, & Zhang, 2015) this sustainability analysis should be conducted for a river basin which is the common spatial unit in water planning process. Blue water footprints relate to the consumptive use of surface and groundwater, whereas green water footprints refer to the use of rainwater that ends up as runoff and does not replenish underground water supplies. Grey water footprint is an indicator of the level of freshwater pollution associated to a stage of a particular process. It is defined as the freshwater volume required assimilating the pollutant load, given the natural background concentration and existing ambient water quality standards. For this study, we will use grey water footprints as the indicator considering

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